

# Machine Learning Parameterizations from the Surface to the Clouds

David John Gagne, Rich Loft, Andrew Gettelman, Jack Chen, Tyler McCandless, Branko Kosovic, Tom Brummet, Sue Ellen Haupt, Bai Yang\*

National Center for Atmospheric Research, \*ERT, Inc., FDR/ARL/NOAA

## Motivation

We use machine learning to either emulate or improve on existing atmospheric parameterization.

1. Microphysics: Can we emulate a computationally-intense scheme within a more efficient scheme?
2. Surface Layer: Can we use many observations to train a ML model to improve on existing schemes?

## Microphysics Methods

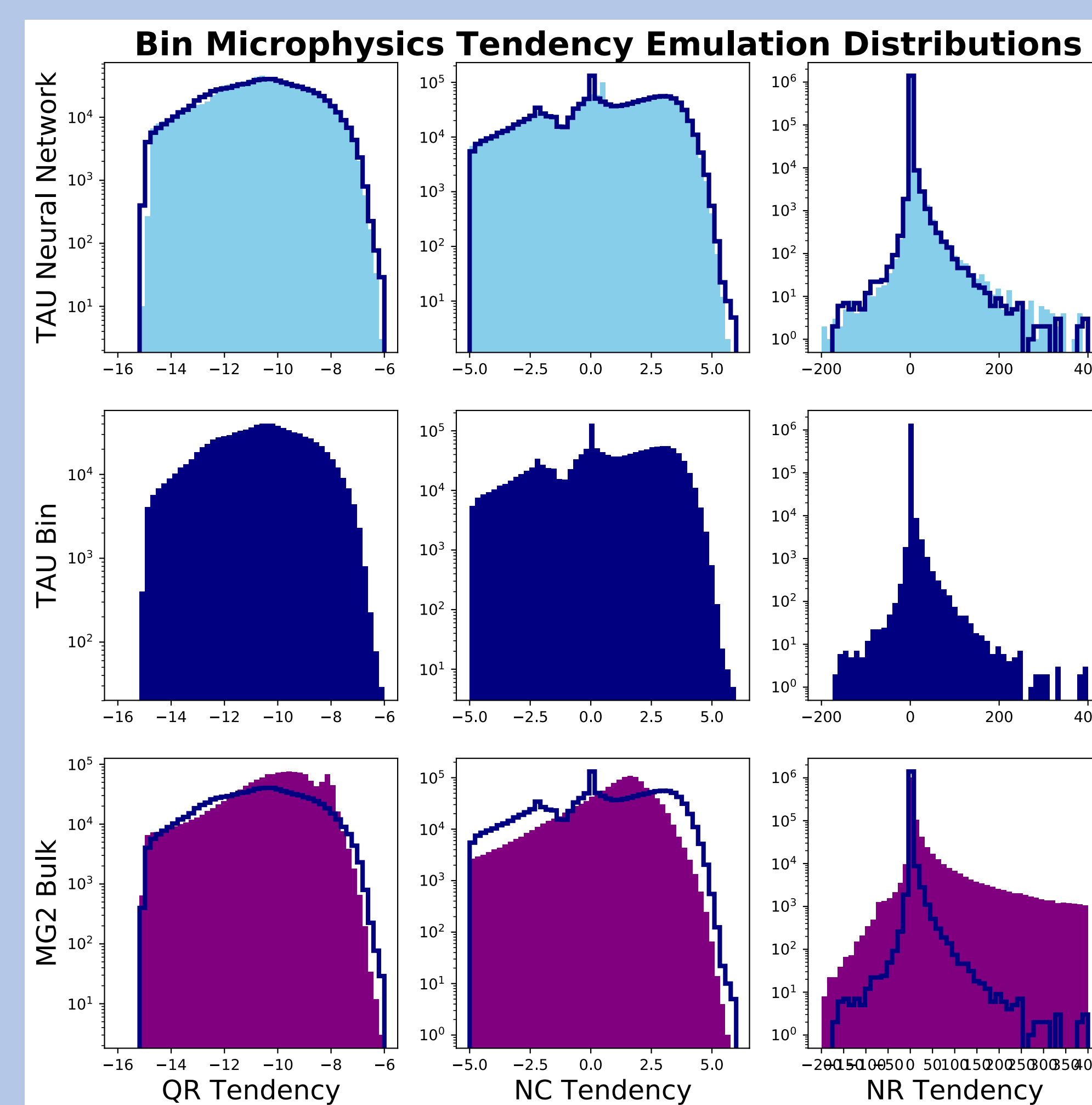
**Problem:** Use neural networks to emulate bin microphysics processes in bulk microphysics scheme

**Data:** Run CAM6 globally for two years globally with TAU bin and MG2 bulk microphysics

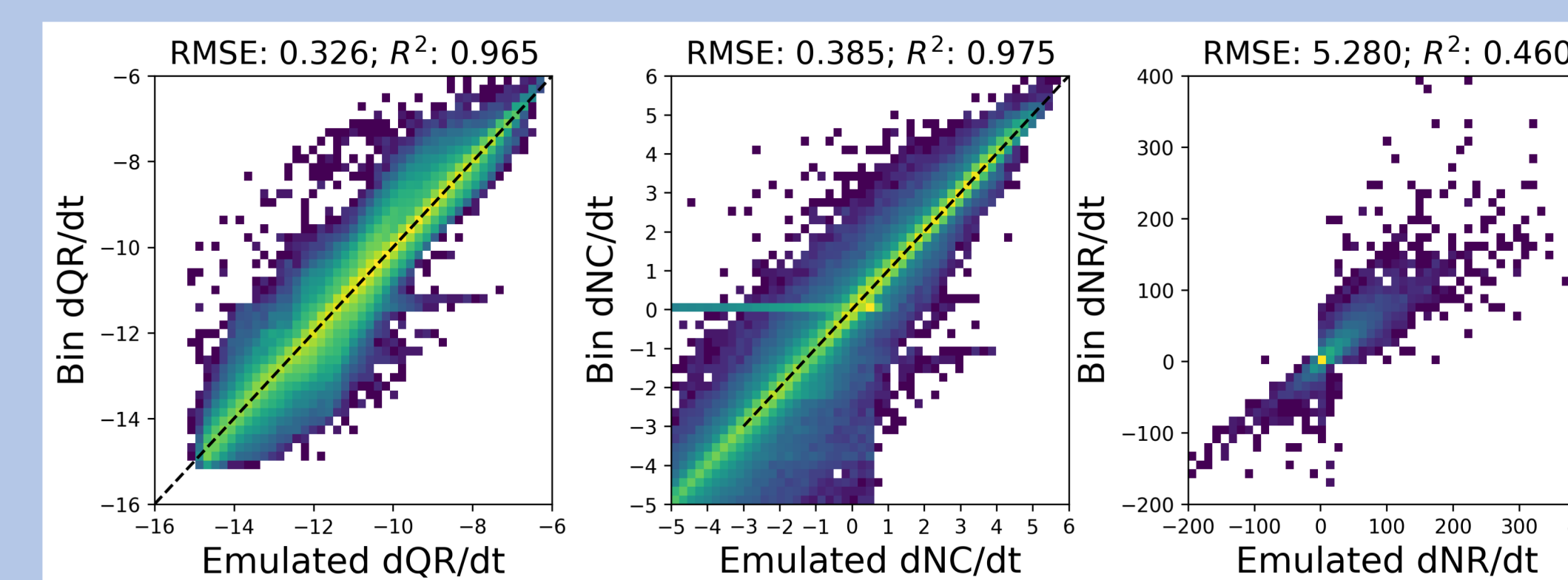
**Inputs:** QC, QR, NC, NR, density  
**Outputs:** Cloud and Rain Water Tendencies

**Models:** Dense neural networks

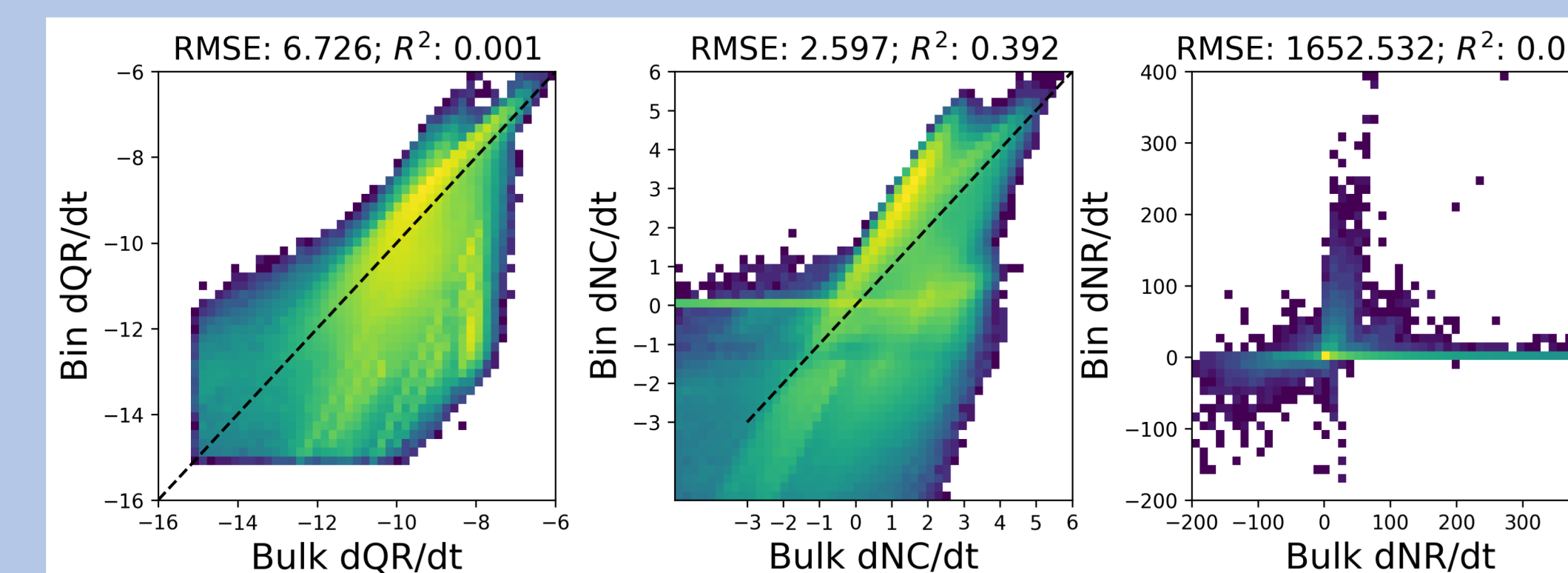
## Neural networks can emulate the distributions and sensitivities of the cloud-to-rain conversion processes in a bin microphysics scheme.



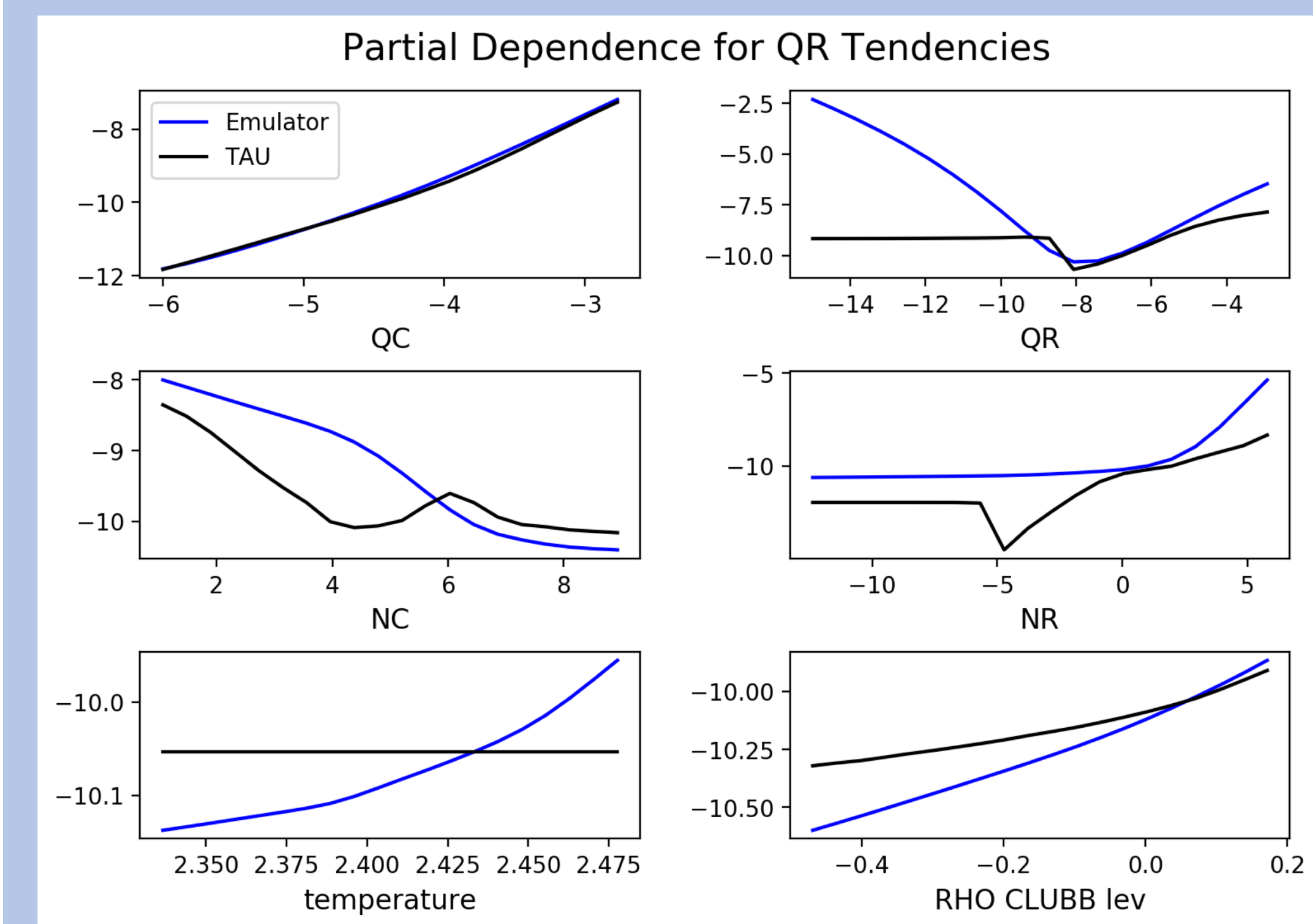
## Neural Network vs. TAU Bin



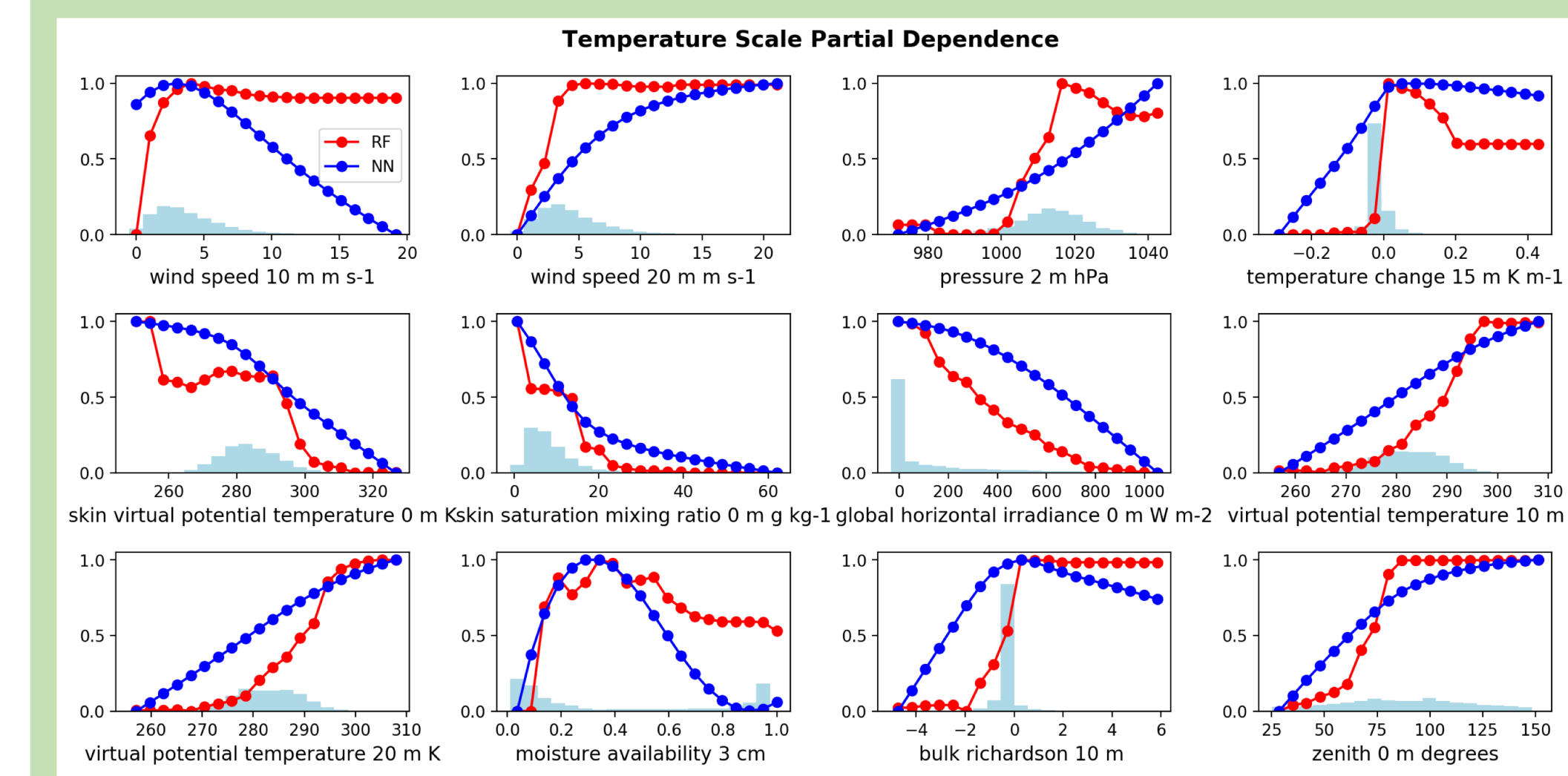
## MG2 Bulk vs. TAU Bin



## Microphysics Partial Dependence



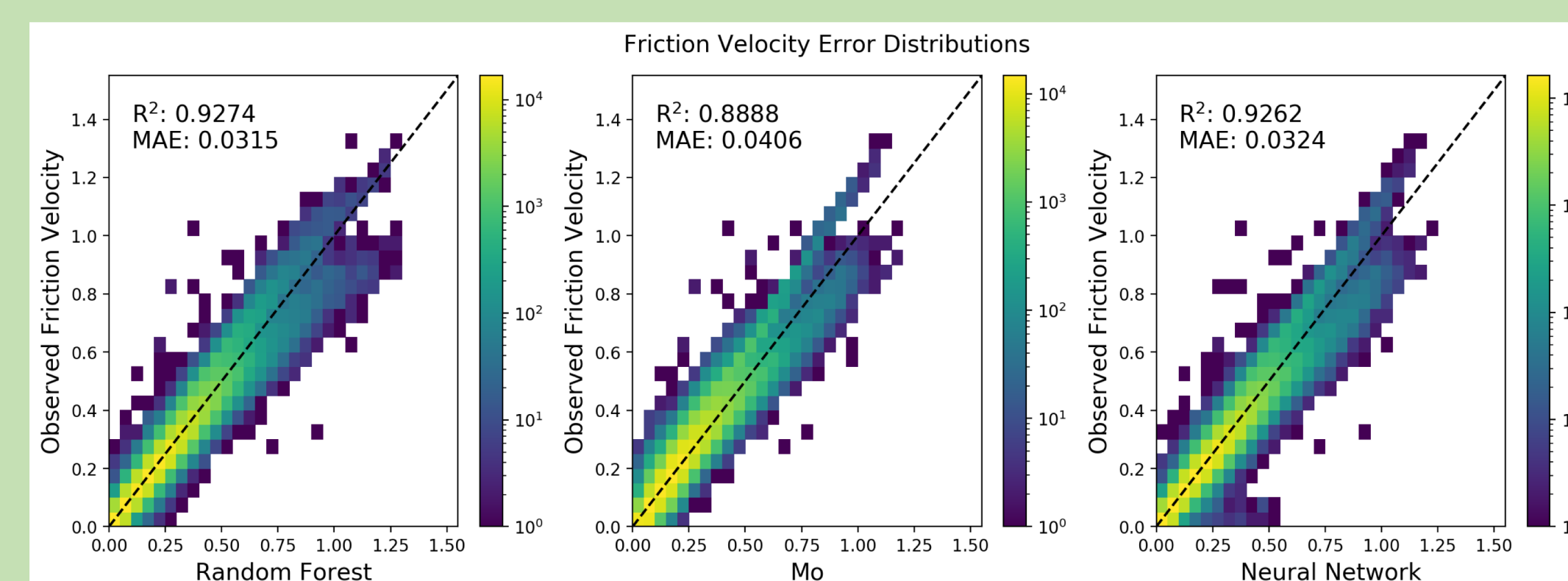
## Surface Layer Partial Dependence



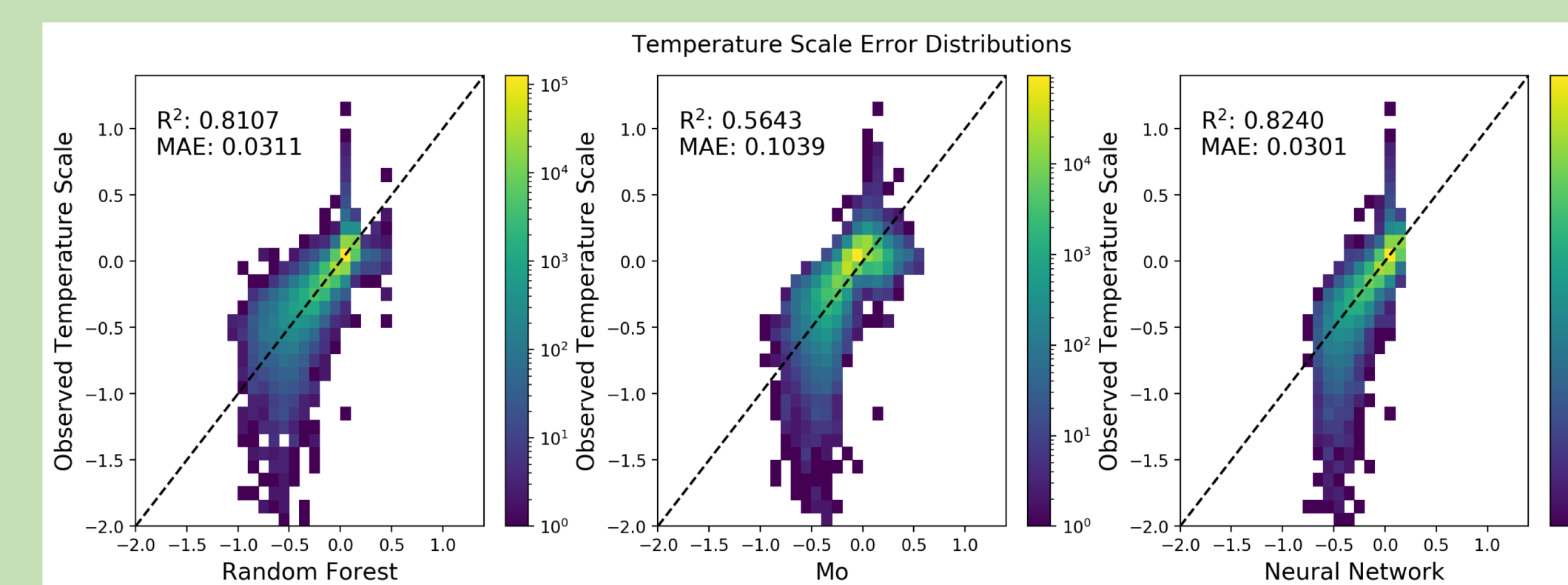
Partial dependence shows the sensitivity of mean prediction to changes in input values.

## Machine learning of surface layer energy fluxes improves on Monin-Obukhov similarity theory.

### Friction Velocity



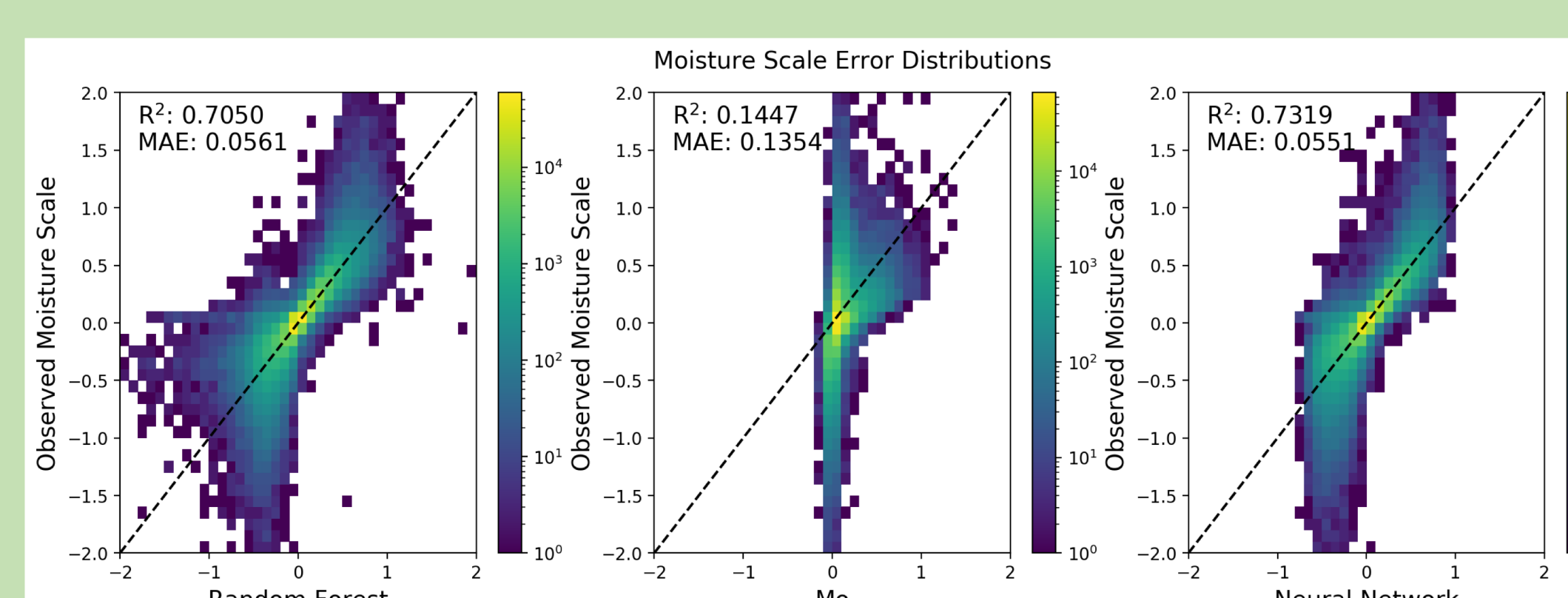
### Temperature Scale



## ML Cross-Testing R<sup>2</sup>

Cabauw Data	U*	$\theta^*$	Q*
MO	0.90	0.44	0.14
RF from Cabauw	0.93	0.82	0.73
RF from Idaho	0.90	0.77	0.49

### Moisture Scale



## Surface Layer Methods

**Problem:** Predict surface momentum and heating fluxes from surface atmospheric profiles

**Data:** Flux and tower observations from Cabauw, Netherlands, and Scoville, Idaho

**Models:** Random forests and neural networks for  $u^*$ ,  $\theta^*$ , and  $q^*$

## Next Steps

Fortran neural network and random forest parameterization modules have been developed. Now integrating Fortran ML parameterizations with WRF and CESM.

## Please Contact Me

Email: [dgagne@ucar.edu](mailto:dgagne@ucar.edu)

Twitter: @DJGagneDos

Github: [djgagne](https://github.com/djgagne)

